## Update on NOAA CO<sub>2</sub> Retrievals: Validation and Future Directions

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NOAA/NESDIS/STAR

#### **AIRS Science Team Meeting**

October 10, 2007

#### **Outline**

- Description of NOAA AIRS CO<sub>2</sub> retrieval methodologies
- How well can we do with a simple climatology?

$$\mathsf{CO}_2(t) = a_0 + a_1 * t$$

- Development of error estimates.
- Paper on averaging kernels (related to these error estimates) accepted (with revisions) to IEEE TGARS.
- Validation with full resolution data vs. NOAA ESRL/GMD Aircraft (2005)¹and Global Gridded data vs. JAL Matsueda (August 2003 -2006.)
- Comparison of AIRS and models What new information can AIRS provide to modeling community?



<sup>&</sup>lt;sup>1</sup>Submitted to JGR in review

- Use AIRS Science Team Methodology.
  - Version 4.7.
    - before cloudy regression introduced
    - NOAA O₂ regression on
  - 70 channels (mostly 15 micron).
  - Derive CO<sub>2</sub> in 4 layers in troposphere, 1 stratospheric.
- Use Optimal Estimation w/ SVD
  - Runs within offline science code (consistent RTA/channel set)
  - Derive 6 10 CO<sub>2</sub> basis functions.
  - Runs very fast No appreciable difference in run-time compared to AIRS Science Team methodology
- Validation with full resolution data vs. NOAA ESRL/GMD Aircraft (2005) and JAL Matsueda between August 2003 - 2006 (only AIRS Science Team approach).

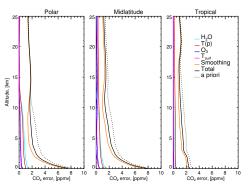
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### Improvement Over Simple Climatology

Theoretical error analysis for our Version 5 Climatology

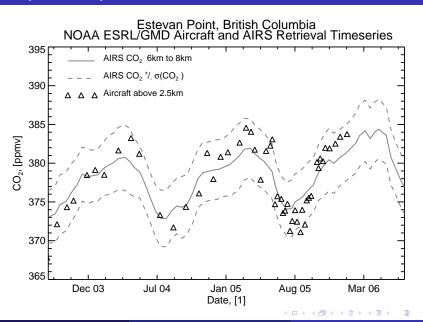


- Calculation uses a priori covariance calculated as the difference between ESRL aircraft and our simple Version 5 climatology.
- Ability to partition error sources and their effect on the retrieval. Effect minimized as we have assumed a *perfect* knowledge of the error covariation of intefering species (assumed *ad-hoc*:  $S(z,z') = \sigma(z)\sigma(z') \cdot \exp(-|z-z'|/L)$ ).

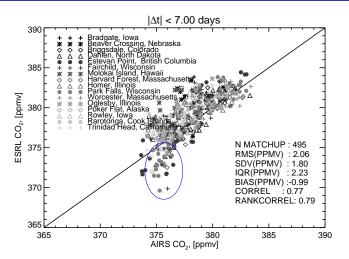
## ESRL/GMD Aircraft Validation Approach

- Use full resolution AIRS retrievals (previously validated w/ 3°x 3°grids)
- Average AIRS CO<sub>2</sub> between 6-10 km (nominally where jacobian has maximum sensitivity).
- Use nominal jacobians (wrt. latitude) to weight ESRL aircraft.
  - Enables comparison of scalar measurements
  - Removes variability in lowest 2.5 km
- Average all retrievals within 200km with temporal matchup window between 1 day - 1 month.
- Profile statistics will also be shown. NOAA/ESRL CarbonTracker<sup>2</sup>model used to extend profiles above 8 km.

### Example Comparison: Estevan Point, British Columbia

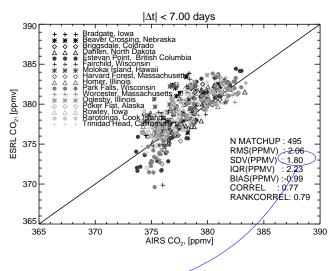


### AIRS Science Team Algorithm vs. ESRL/GMD Aircraft



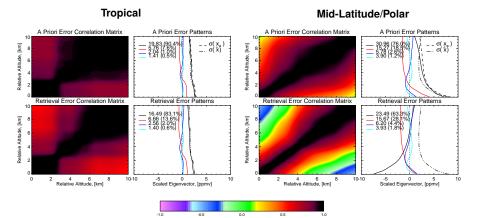
 Total magnitude of drawdown at LEF not captured possible over-regularization wrt. characteristic variability.

### AIRS Science Team Algorithm vs. ESRL/GMD Aircraft



0.5% uncertainty from space!

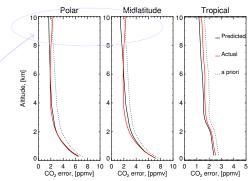
### Calculated a priori and Retrieval Error Covariances



- Total variance of the retrieval is less than the a priori indicating a gain in information.
- First eigenfunction variance (and percent of total variance) of the retrieval is less than a priori.
- Retrieval tends to redistribute variance among higher order eigenfunctions, which are similar in shape to the *a priori*, indicating we have only 1 piece of information, albeit well constrained, in the vertical. Vertical resolution ≈6-8 km.

### Validation of Error Propagation

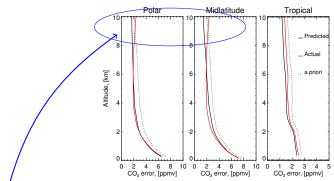
In general, predicted errors and actual errors compare very well.



- Largest discrepancy is above 8 km where the NOAA CarbonTracker model was used to extend the aircraft profiles.
- Uncertainties in the profile extension procedure, the model profiles, AIRS retrievals and/or error analysis are possible explanations to the disagreement.

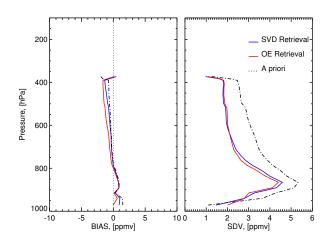
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## Comparison of OE and SVD approaches: independent validation

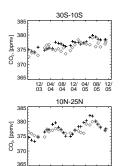


 Two retrievals with completely different methods of regularization yield almost the same results.

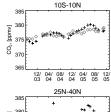
## JAL Aircraft Validation Approach

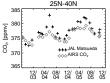
- NOAA 3°x 3°gridded subset
- Average AIRS CO<sub>2</sub> between 6-10 km (nominally where jacobian has maximum sensitivity).
- Average all retrievals within 1000km with temporal matchup 1 month.
- Compare to monthly averaged JAL Matsueda over latitude range (27 months total between August 2003-2006).

### AIRS Science Team Algorithm vs. JAL Matsueda



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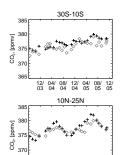




- SDVE < 1.5 ppmv for all latitude ranges</li>
- Variability in the accuracy wrt. latitude on the order of retrieval precision
  - related to sensitivity of jacobians to H<sub>2</sub>O displacement.
  - zonal variability of information content.
- Averaged over all latitudes, AIRS retrievals compare very well:
  - -0.62  $\pm$  0.87 ppmv

Latitude	SDVE	BIAS
Range	[ppmv]	[ppmv]
30S - 10S	1.32	-1.08
10S - 10N	1.04	-0.06
10N - 25N	1.45	-0.42
25N - 40N	1.45	-1.43
30S - 40N	0.87	-0.62

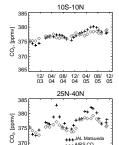
### AIRS Science Team Algorithm vs. JAL Matsueda



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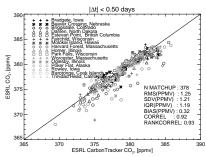
365

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Latitude SDVE **BIAS** Range [ppmv] [ppmv] 30S - 10S 1.32 -1.08 10S - 10N 1.04 -0.0610N - 25N 1.45 -0.4225N - 40N 1.45 -1.4330S - 40N 0.87 -0.62

## NOAA ESRL/GMD CarbonTracker vs ESRL/GMD Aircraft

NOAA ESRL/GMD CarbonTracker weighted using AIRS jacobians.

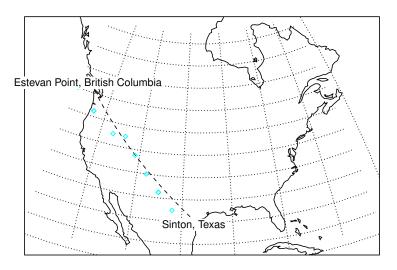


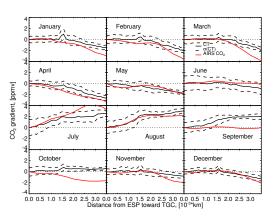
- 0.5 ppmv better precision than AIRS baseline, however CarbonTracker has been optimized for N. America.
- From our eigenvector analysis of our a priori, the 1<sup>st</sup> eigenfunction, a total column perturbation, explains 80-90% of the variance.
- We would expect good agreement near ESRL aircraft sites because constraint of having surface / tower measurements in the assimilation.

### Approach to Estimate AIRS Impact

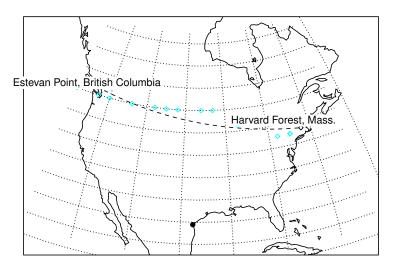
- Determine scales of variability in CarbonTracker calculated as the gradient in a given direction over a defined time scale.
- Compare to see if AIRS captures the same sort of gradients.
- 3°x 3°grids used for comparison.

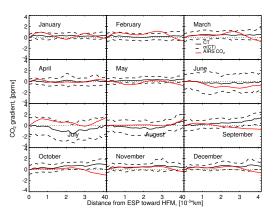
Thanks to Wouter Peters (NOAA ESRL) for suggesting using CarbonTracker for this approach.





- 1-σ monthly variability of FT gradients shows that in general we need to resolve 1 ppmv signals (larger variability in summer months due to rectifier) on short timescales.
- Monthly averaged free tropospheric (FT) gradients are within our expected error budget in terms of matching seasonality and horizontal placement.





- CarbonTracker shows lack of FT gradient due to rapid advection/mixing of surface fluxes.
- East-to-west 1-σ variability largest in the summer months due to frontal passages and hence strong mixing (weekly differences in gradients ≈ ±3 ppm).
- Considering retrieval error budget (wrt. aircraft) we may be able to resolve these features on weekly timescales; however, more study is required.

### Summary

- Able to provide global retrievals of CO<sub>2</sub> on 1-2 weekly timescales at 1 - 2ppmv precision with a globally fixed a priori.
  - Modeling groups at NASA/GSFC, UC/Berkeley, and University of Leicester, UK have just begun looking at the product.
- Theoretical error estimates enable quick calculation of the AIRS data impact. These require accurate large scale correlations in a priori due to the broad width of the kernel functions.
- Require more high altitude profile validation data to gain confidence in product error correlation.

### Summary: Future Plans

- True test of product skill is the ability to discern CO<sub>2</sub> gradients.
- Model gradients W-E are generally small due to rapid advection of surface fluxes – we may be able to capture weekly differences.
- As expected N-S gradients are larger with monthly variability on the order of our precision.
- Monthly comparisons to CarbonTracker show similar features; more analysis required
  - Determine our ability to match gradients over shorter timescales.
  - Retest AIRS in regions poorly constrained. Model/retrieval comparisons underway for gradient appropriateness.
  - $\begin{tabular}{ll} \hline \textbf{0} & Understand (inter) product error correlations $f(time,space)$ that introduce anomalous gradients in AIRS $CO_2$. \\ \hline \end{tabular}$

### Theoretical Gain Using a CarbonTracker a priori

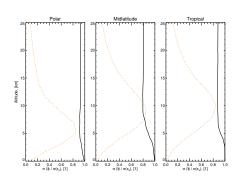
• Use error propagation to estimate gain in information content by adding AIRS  $\mathrm{CO}_2$  sensitive measurements initially with CarbonTracker errors,  $\mathbf{S}_a^{\mathrm{CTracker}}$ .

$$\begin{split} \mathbf{\hat{S}} &= (\mathbf{A} - \mathbf{I}) \mathbf{S}_a^{\mathsf{CTracker}} (\mathbf{A} - \mathbf{I})^T \\ &+ \mathbf{D} \mathbf{K_b} \mathbf{S_b} (\mathbf{D} \mathbf{K_b})^T \\ &+ \dots \end{split}$$

We plot the error reduction defined as:

$$diag(\hat{S})/diag(S_a^{CTracker})$$

 Improvement outside of region where Jacobian (dotted line) is sensitive is largely due to error correlation assumed in S<sub>a</sub><sup>CTracker</sup>.



Improvement somewhat marginal; however, CarbonTracker is highly constrained by surface measurements hence  $\mathbf{S}_a^{\text{CTracker}}$  is small.